引用時請先將前面的灰底部分刪除

註：以下依照演算法發表時間排序，但文章正式發表時間和演算法發表順序不同，所以可能出現有些編號前面的文章但引用年分較晚的情形。

其中引用為arXiv的為尚未正式發表的文章，所以引用也非正式引用。待該文章有正式引用格式後，再行補上。

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